

Modeling Health Insurance Coverage Estimates for Minnesota Counties

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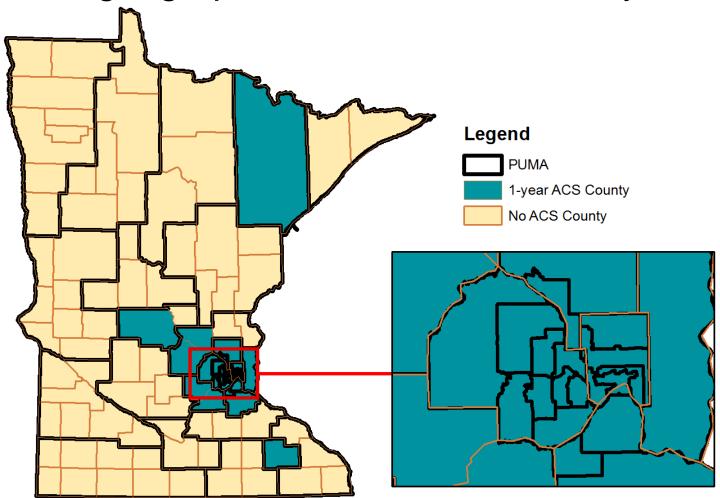
Background

- Minnesota Health Access Survey (MNHA)
 - Telephone survey conducted every 2 years
 - Provides MN and regional estimates, including estimates for select populous counties and cities
 - County level estimates are frequently requested data
- American Community Survey (ACS)
 - Estimates available for all PUMAs (Public Use Microdata Areas)
 - Estimates available for 12 Minnesota counties (out of 87)
- Small Area Health Insurance Estimates Program (SAHIE)
 - 2007 estimates for all 87 Minnesota counties
 - 2009 estimates are now available but not examined in this edition of the model



Background

Minnesota geographies and data availability



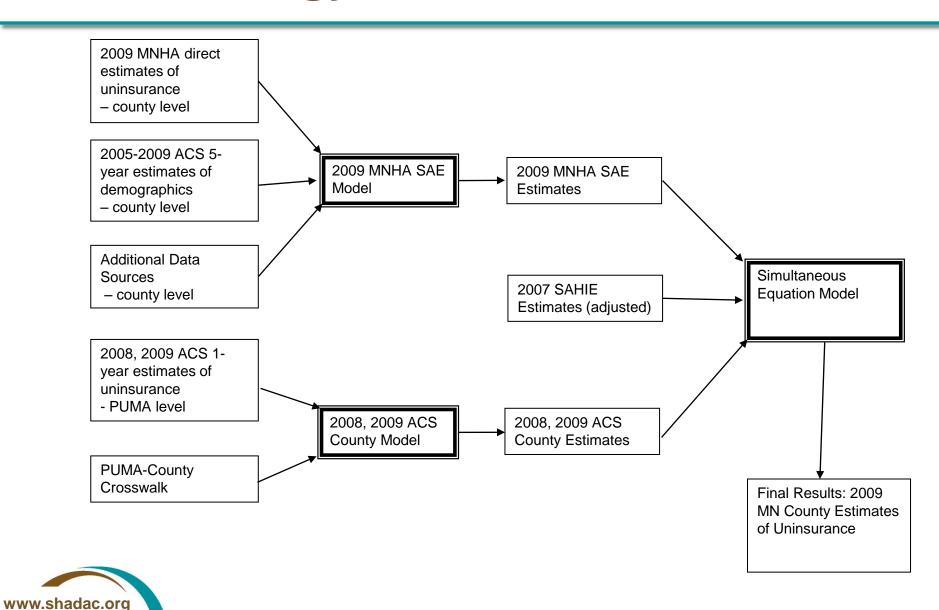


Research Objective

- Produce Minnesota uninsurance rates by county for 2009
 - Use the Minnesota Health Access Survey (MNHA)
 - Use other sources of uninsurance estimates
 - Include estimates of uncertainty
 - Allow for future input sources
 - Create methodology that is incrementally observable
 - Use methods that can be applied to other states



Methodology Overview



MNHA SAE: Model

Estimates come from normal distribution

$$y_c^{MNHA-direct} \sim N\left(\overline{y_c}, \frac{1}{\tau}\right)$$

Model the mean using covariates X and error

$$\overline{y_c} = \alpha + \beta X + v_c$$

Error is correlated spatially with neighbors

$$v_c|v_{-c}, \sigma_v^2 \sim N\left(\sum_{j \in \delta_c} \frac{v_j}{|\delta_c|}, \frac{\sigma_v^2}{|\delta_c|}\right)$$



MNHA SAE: Model Parameters

<u>Parameters</u>	<u>Prior</u>	<u>Median</u>	<u>SE</u>
Percent Moved into State, 2005-2009	$N(0,1/1\times10^6)$	1.501	0.6422
Percent White, 2005-2009	$N(0,1/1\times10^6)$	-0.3024	0.1143
Percent HHLDS 65 and Over, 2005-2009	$N(0,1/1\times10^6)$	0.2638	0.1089
Percent of Population Growth, 2000-2009	$N(0,1/1\times10^6)$	-2.643	0.9837
Percent Land in Farms, 2007	$N(0,1/1\times10^6)$	0.05293	0.02414
Percent Employed Working in Retail, 2009	$N(0,1/1\times10^6)$	0.5771	0.244
Average Unemployment Rate, 2009	$N(0,1/1\times10^6)$	2.102	0.3965
Weekly Wage, 2009	$N(0,1/1\times10^6)$	0.02775	0.007752
Constant	$N(0,1/1\times10^6)$	-16.63	13.36
Precision v	$\Gamma(0.001, 0.001)$	2.262	94.55
Precision $ au$	$\Gamma(0.001, 0.001)$	0.03686	0.006257
	DIC	541.9	

Pd



7.788

ACS County Model

- County estimate from 1-year ACS (12 counties)
 - Estimate and SE used directly
- County is a subset of PUMA (75 counties)
 - Use the relationship between puma and county for the poverty rate to estimate the county given a puma uninsurance rate using equations 1-3
 - 1) $Unin_c^{puma} = \beta_0 + \beta_1 Pov_c^{puma}$; c = 1, 2, ..., 87
 - 2) $Pov_diff_c^{puma} = Pov_c^{County} Pov_c^{puma}$
 - 3) $unin_c^{county} = Unin_c^{puma} + \beta_1 Pov_diff_c^{puma}$
 - SE is the PUMA estimate times the ratio of the PUMA poverty SE divided by the county poverty SE

$$unin_se_c^{county} = unin_se_c^{puma} \sqrt{\left(\frac{pov_se_c^{county}}{pov_se_c^{puma}}\right)}$$



SAHIE Estimate & Adjustment

- Census Bureau's Small Area Health Insurance Estimates (SAHIE) program produces modelbased estimates of health insurance coverage
- Estimates are for 0-64 so we need to make a correction to use in our all ages model

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\begin{aligned} Unin_{All}^{SAHIE} &= Unin_{under65}^{SAHIE} - \left(Unin_{under65}^{SAHIE} * Unin_{under65}^{SAHIE}\right) \\ &+ \left(Prop65over^{ACS5year} * Unin_{65over}^{CPS}\right) \end{aligned}
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Simultaneous Equation Model (SEM)

Each survey-county estimate from a normal distribution

$$y_{sc} \sim N(u_{sc}, \tau_{sc}^u)$$

 The precision is a survey term times a surveycounty specific error

$$\tau_{sc}^u = \tau_s * \tau_{sc}^\tau$$

The survey-county error is the inverse of the estimate's variance

$$\tau_{SC}^{\tau} = \frac{1}{\sigma_{SC}^2}$$



Simultaneous Equation Model (SEM)

$$u_c^{MNHA_SAE_2009} = \alpha_1 + \beta_c \ County_c$$
 $u_c^{ACS_2009} = \alpha_2 + \beta_c \ County_c$
 $u_c^{ACS_2008} = \alpha_3 + \beta_c \ County_c$
 $u_c^{SAHIE_2007} = \alpha_4 + \beta_c \ County_c$

$$y_c^{SEM} = (\alpha_1 + \alpha_2 + \alpha_3 + \alpha_4)/4 + \beta_c \ County_c \longrightarrow \text{Prediction}$$

Specifications

- Single Markov Chain Monte Carlo
- 20k production cycles after 1,000 burn-in iterations



SEM: Model Parameters

<u>Parameters</u>		<u>Prior</u>	<u>Median</u>	<u>SD</u>
$lpha_{MNHA_SAE_2009}$		$N(0,1/1\times10^6)$	30.49	8.282
$lpha_{ACS_2009}$		$N(0,1/1\times10^6)$	29.77	8.282
$lpha_{ACS_2008}$		$N(0,1/1\times10^6)$	29.4	8.283
$lpha_{SAHIE_2007}$		$N(0,1/1\times10^6)$	29.7	8.281
eta_{1-87}		$N(0,1/1\times 10^6)$	-23.9415.21	8.29-8.428
τ		$\Gamma(0.001, 0.001)$	0.5335	0.04714
	DIC		1558	
	Pd		91.27	

SD: Standard Deviation

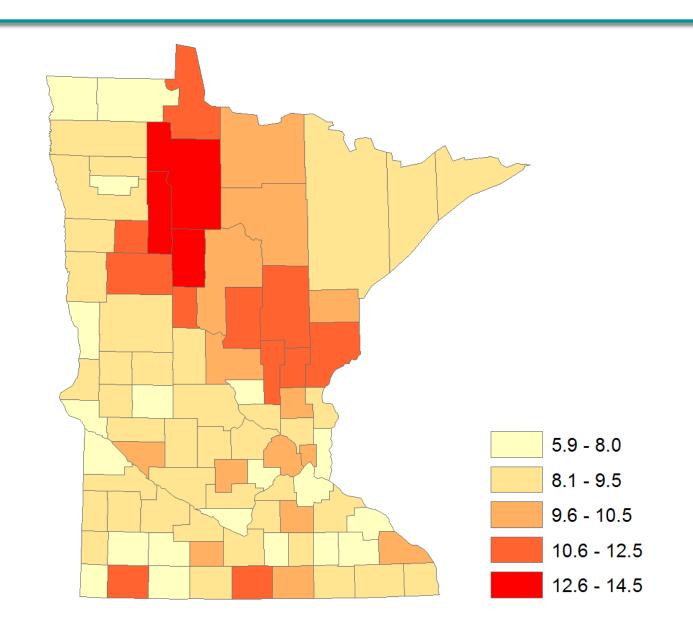


Methodology Limitations/Enhancements

- MNHA SAE model could include more advanced variable selection and transformations of covariates
- MNHA SAE model could take advantage of information outside the state (eg. US counties)
- Assumptions about PUMA to county relationships for ACS are not currently testable
- SEM Model excludes non-parametric errors
 - Integrated model could propagate errors more accurately but sacrifice conceptual simplicity

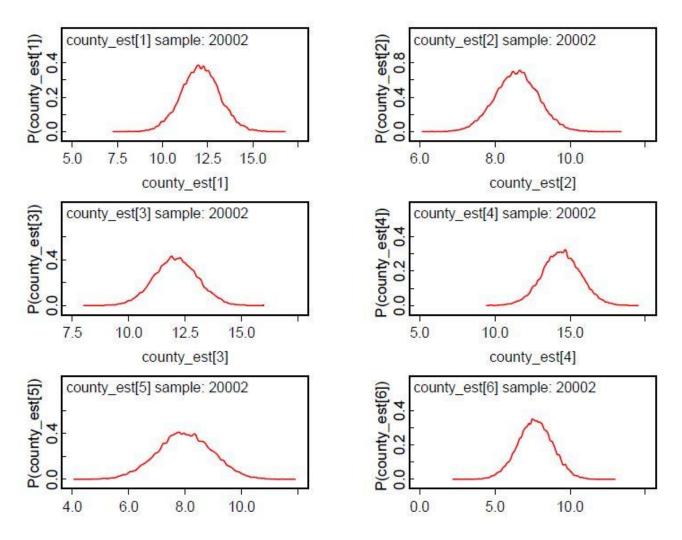


Model Results - Percent Uninsured





Model Results: Posterior Density

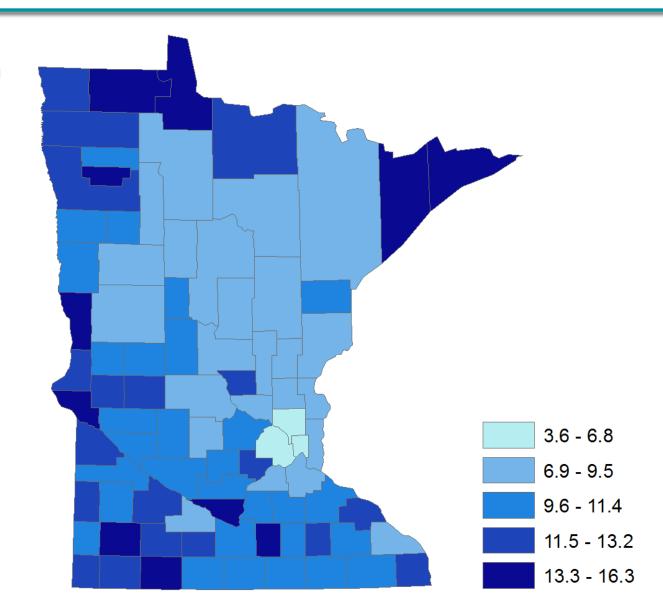




Model Results - Uncertainty

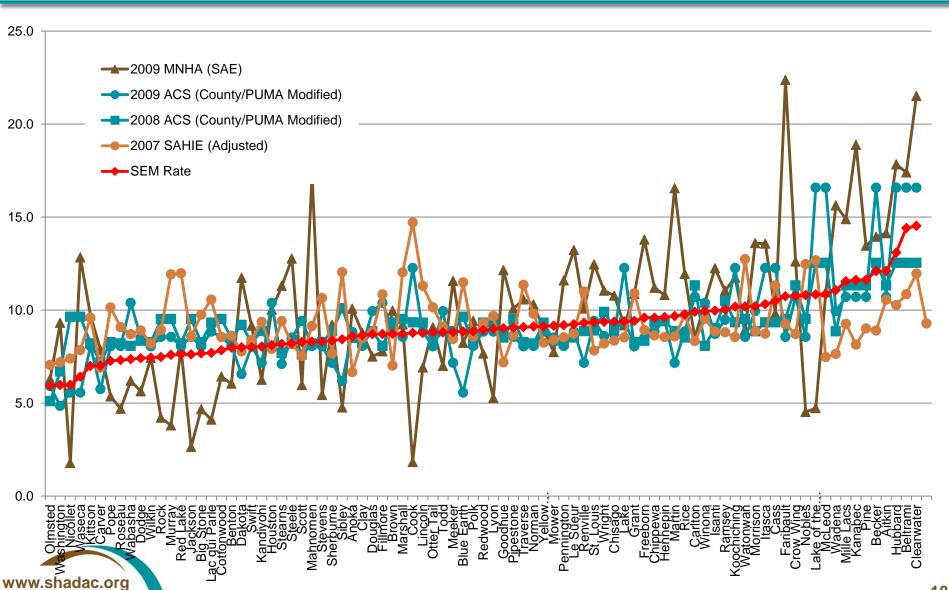
Coefficient of Variation

$$CV = \frac{SD}{EST}$$





Model Results: Input Comparison



Conclusion

- Produced uninsurance estimates and estimates of uncertainty using a state survey and multiple input sources
- Methodology is accessible and can be applied to other states and new input sources
- Results are important for states who need to prepare for changes under health reform

